

HW #4

① Training a neural network. (NN)

generate $N=100$ equidistant data points using:

(a) $f(x) = x^2 + \xi$, $x \in [-5, 5]$

(b) $f(x) = \sin(x) + \xi$, $x \in [0, 2\pi]$

Here, $\xi = \mathcal{N}(\underbrace{0}_{\mu}, \underbrace{0.05^2}_{\sigma^2})$ is random noise.

Implement a two-layer feed-forward NN with $M=5$ hidden units. Choose the appropriate input & output units and justify this choice, with the goal of fitting $f(x)$ by regression.

Sketch the NN architecture and

(or plot) train the NN by error backpropagation

coupled with stochastic gradient descent.

Plot the data and the fit for both

(a) & (b).

② Bishop 5.26

③ Implement a binary Hopfield network (HN) with $I = 60$ spins. Choose 2 distinct memories \vec{x}_1 & \vec{x}_2 & compute w_{ij} by the Hebbian rule:

$$w_{ij} = \sum_{n=1}^2 x_i^{(n)} x_j^{(n)}$$

Each spin can adopt ± 1 values.

(a) Start $N=100$ runs from randomly chosen spin configurations and find the local energy minima of the energy function:

$$E = -\frac{1}{2} \sum_{i,j} w_{ij} x_i x_j \quad [\text{no biases}]$$

by asynchronous updates (cycle through all spins one at a time).

what are the # of times memories 1 and 2 were recovered?

(b) Now, set a quarter of all weights w_{ij} to \emptyset randomly.

Repeat (a) starting from the same initial spin configurations & compare the # of times memories 1 and 2 were recovered with the results of (a).